

An Alternative Derivation of Johannisson's Regular Perturbation Model

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Abstract

We provide here an alternative derivation of the generalization of the nonlinear Turin model for dispersion unmanaged coherent optical links provided in Johannisson's report [1].

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I. INTRODUCTION

Goal of this paper is to provide a simplified derivation of the results appearing in the recent ArXiv posting of P. Johannisson [1] on a generalization of the well-known frequency-domain nonlinear interference (NLI) analytical model for dispersion unmanaged (DU) coherent systems introduced by Turin's group in [2].

II. FREQUENCY DOMAIN NLI RP1 SOLUTION

We start from the dual-polarization (DP) single-channel first-order Regular Perturbation (RP1) solution of the dispersion-managed nonlinear Schroedinger equation (DMNLSE) ([3], Appendix 2):

$$\tilde{U}(L, f) = \tilde{U}(0, f) + \tilde{U}_p(L, f) \quad (1)$$

where L is the total link length, and the NLI perturbation field is

$$\tilde{U}_p(L, f) = -jP_0 \iint_{-\infty}^{\infty} \mathcal{K}(f_1 f_2) \tilde{U}(0, f + f_1) \tilde{U}^\dagger(0, f + f_1 + f_2) \tilde{U}(0, f + f_2) f_1 f_2 \quad (2)$$

where:

i) boldface fields are 2x1 vectors containing the Fourier transforms (denoted by a tilde) of the X and Y polarizations in the transmitter polarization frame of reference; a dagger stands for transposition and conjugation; and the DP field power is normalized to an arbitrary reference power P_0 (which in [1] is chosen as the *per-polarization* average power);

ii) the un-normalized scalar frequency kernel is defined as (Cfr. [1] eq. (45) and [3] eq. (25)):

$$\mathcal{K}(F) \triangleq \int_0^L \gamma'(s) \mathcal{G}(s) e^{-jC(s)(2\pi)^2 F} ds \quad (3)$$

where $F = f_1 f_2$ is the product of two frequencies, $\gamma' = \frac{8}{9}\gamma$ with γ the fiber nonlinear coefficient, $\mathcal{G}(s)$ the line power gain from $z = 0$ to $z = s$, and $C(s) = -\int_0^s \beta_2(z) dz$ is the cumulated dispersion in the transmission fibers (with dispersion coefficient β_2) up to coordinate s . We choose here the frequency-normalizing rate in [3] as $R = 1$, i.e., we do not normalize the frequency axis, as in [1]. In [3] we use the normalized kernel

$$\tilde{\eta}(F) = \frac{\mathcal{K}(F)}{\mathcal{K}(0)}$$

which then at $F = 0$ equals 1. The nonlinear phase referred to power P_0 is

$$\Phi_{NL} = P_0 \mathcal{K}(0).$$

Hence by multiplying and dividing by $\mathcal{K}(0)$ we can recast (2) as

$$\tilde{U}_p(L, f) = -j\Phi_{NL} \iint_{-\infty}^{\infty} \tilde{\eta}(f_1 f_2) \tilde{U}(0, f + f_1) \tilde{U}^\dagger(0, f + f_1 + f_2) \tilde{U}(0, f + f_2) f_1 f_2 \quad (4)$$

From (4), the X component of the RP1 solution writes explicitly as

$$\begin{aligned} \frac{\tilde{U}_{x,p}(L, f)}{-j\Phi_{NL}} &= \iint_{-\infty}^{\infty} \tilde{\eta}(f_1 f_2) \tilde{U}_x(0, f + f_1) \tilde{U}_x^*(0, f + f_1 + f_2) \tilde{U}_x(0, f + f_2) f_1 f_2 + \\ &\quad \iint_{-\infty}^{\infty} \tilde{\eta}(f_1 f_2) \tilde{U}_x(0, f + f_1) \tilde{U}_y^*(0, f + f_1 + f_2) \tilde{U}_y(0, f + f_2) f_1 f_2 \end{aligned} \quad (5)$$

where the first line gives the self-phase modulation (SPM) of X on X, while the second line gives the intra-channel cross-polarization modulation (I-XPoM) of Y on X. A perfectly dual expression for component Y is obtained by exchanging the indices x and y .

III. GAUSSIAN ASSUMPTION AND JOHANNISSON'S RESULT

In [1], [2] the key assumption is that the input fields are composed of independent spectral lines with *Gaussian* amplitude:

$$\begin{aligned} \tilde{U}_x(0, f) &= \sqrt{f_0} \sum_{k=-\infty}^{\infty} \xi_k \sqrt{\hat{G}_x(k f_0)} \delta(f - k f_0) \\ \tilde{U}_y(0, f) &= \sqrt{f_0} \sum_{k=-\infty}^{\infty} \zeta_k \sqrt{\hat{G}_y(k f_0)} \delta(f - k f_0) \end{aligned}$$

with ξ_k and ζ_k independent identically distributed standard (i.e. zero-mean unit variance) circular complex Gaussian random variables (RV). Such signals do have a *per-polarization* power spectral density $\hat{G}_{x/y}(f)$ (normalized to P_0) in the limit $f_0 \rightarrow 0$ [2]. Then after long statistical averaging calculations, one gets the power spectral density of the $\tilde{U}_{x,p}(L, f)$ RV as ([1], eq. (89)). Note that our PSD $\hat{G}(f)$ is normalized such that $G(f) \equiv P_0 \hat{G}(f)$, where G is the un-normalized PSD per polarization. Also, $P_x = P_0 \int_{-\infty}^{\infty} \hat{G}_x(f) df$ and $P_y = P_0 \int_{-\infty}^{\infty} \hat{G}_y(f) df$

$$\begin{aligned} \hat{G}_{x,p}(f) &= P_0^2 \{ 2 \iint_{-\infty}^{\infty} |\mathcal{K}((f_1 - f)(f_2 - f))|^2 \hat{G}_x(f_1) \hat{G}_x(f_2) \hat{G}_x(f_1 + f_2 - f) f_1 f_2 \\ &\quad + \iint_{-\infty}^{\infty} |\mathcal{K}((f_1 - f)(f_2 - f))|^2 \hat{G}_x(f_1) \hat{G}_y(f_2) \hat{G}_y(f_1 + f_2 - f) f_1 f_2 \\ &\quad + \mathcal{K}(0)^2 \hat{G}_x(f) \left(4 \left(\int_{-\infty}^{\infty} \hat{G}_x(f) df \right)^2 + 4 \int_{-\infty}^{\infty} \hat{G}_x(f) df \int_{-\infty}^{\infty} \hat{G}_y(f) df + \left(\int_{-\infty}^{\infty} \hat{G}_y(f) df \right)^2 \right) \} \end{aligned} \quad (6)$$

and a dual expression for Y is obtained by swapping $x \leftrightarrow y$. Recall that $\hat{G}_{x,p}(f)$ is the NLI PSD, normalized by P_0 .

An equivalent form of (6) is the following

$$\begin{aligned} \hat{G}_{x,p}(f) &= \Phi_{NL}^2 \{ 2 \iint_{-\infty}^{\infty} |\tilde{\eta}(f_1 f_2)|^2 \hat{G}_x(f + f_1) \hat{G}_x(f + f_2) \hat{G}_x(f + f_1 + f_2) f_1 f_2 \\ &\quad + \iint_{-\infty}^{\infty} |\tilde{\eta}(f_1 f_2)|^2 \hat{G}_x(f + f_1) \hat{G}_y(f + f_2) \hat{G}_y(f + f_1 + f_2) f_1 f_2 \\ &\quad + \hat{G}_x(f) \left(4 \left(\int_{-\infty}^{\infty} \hat{G}_x(f) df \right)^2 + 4 \int_{-\infty}^{\infty} \hat{G}_x(f) df \int_{-\infty}^{\infty} \hat{G}_y(f) df + \left(\int_{-\infty}^{\infty} \hat{G}_y(f) df \right)^2 \right) \} \end{aligned} \quad (7)$$

which better shows the formal parallel with the field equation (5): the field double integral in f_1, f_2 of the product kernel-field-field*field becomes a PSD double integral in f_1, f_2 of the product squared kernel magnitude-PSD-PSD-PSD.

It is the purpose of the remaining part of this paper to provide a new proof of (7).

IV. THE NEW PROOF

We now start from (5) and make the following two assumptions regarding the input X,Y fields $U_x(0, t), U_y(0, t)$:

- 1) they are wide-sense stationary (WSS);
- 2) they are jointly Gaussian processes.

Regarding assumption 1), we plan to exploit the following extension of result ([4], p. 418, eq. (12-76)):

Theorem 1

Consider the jointly WSS stochastic processes $x(t)$ and $y(t)$, and let

$$\tilde{X}(f) \equiv \mathcal{F}[x(t)] = \int_{-\infty}^{\infty} x(t) e^{-j2\pi f t} dt$$

the Fourier transform of x (in the mean-square (MS) sense), and $\tilde{Y}(f)$ is similarly defined. Let their cross power spectral density (PSD) be $G_{xy}(f) = \mathcal{F}[R_{xy}(\tau)] = \mathcal{F}[E[x(t + \tau)y^*(t)]]$. Then

$$E[\tilde{X}(f)\tilde{Y}^*(u)] = G_{xy}(f)\delta(f - u) \equiv G_{xy}(f)\delta(u - f) \quad \square \quad (8)$$

As a byproduct, we also have

$$E[\tilde{X}(f)\tilde{X}^*(u)] = G_x(f)\delta(f - u) \equiv G_x(f)\delta(u - f).$$

This theorem thus shows that the Fourier transform of *any MS-integrable WSS process* is nonstationary white noise, and thus *the spectral lines of its Fourier transform are uncorrelated*.

Regarding assumption 2), we plan to exploit the following result, known as the *complex Gaussian moment theorem* (CGMT), a generalization to complex variables of Isserlis theorem [5], [6]:

Theorem 2

Let U_1, U_2, \dots, U_{2k} be zero-mean jointly circular complex Gaussian random variables. Then

$$E[U_1^* U_2^* \dots U_k^* U_{k+1} U_{k+2} \dots U_{2k}] = \sum_{\pi} E[U_1^* U_p] E[U_2^* U_q] \dots E[U_k^* U_r] \quad (9)$$

where \sum_{π} denotes a summation over the $k!$ possible permutations (p, q, \dots, r) of indices $(k+1, k+2, \dots, 2k)$ \square

For instance,

$$\begin{aligned} E[U_1^* U_2^* U_3^* U_4 U_5 U_6] &= E[U_1^* U_4] E[U_2^* U_5] E[U_3^* U_6] \\ &+ E[U_1^* U_4] E[U_2^* U_6] E[U_3^* U_5] \\ &+ E[U_1^* U_5] E[U_2^* U_4] E[U_3^* U_6] \\ &+ E[U_1^* U_5] E[U_2^* U_6] E[U_3^* U_4] \\ &+ E[U_1^* U_6] E[U_2^* U_4] E[U_3^* U_5] \\ &+ E[U_1^* U_6] E[U_2^* U_5] E[U_3^* U_4]. \end{aligned} \quad (10)$$

Let's now start the new proof. We are interested in the PSD $\hat{G}_{x,p}(f)$ of the NLI field $U_{x,p}(L, t) = \mathcal{F}^{-1}[\tilde{U}_{x,p}(L, f)]$. By theorem 1 we have:

$$E[\tilde{U}_{x,p}(L, f)\tilde{U}_{x,p}^*(L, u)] = \hat{G}_{x,p}(f)\delta(u - f). \quad (11)$$

The left hand side can be explicitly calculated using (5):

$$\begin{aligned}
& \frac{E[\tilde{U}_{x,p}(L, f)\tilde{U}_{x,p}^*(L, u)]}{\Phi_{NL}^2} = \\
& E\left[\int_{-\infty}^{\infty}\tilde{\eta}(f_1f_2)[\tilde{U}_x(0, f+f_1)\tilde{U}_x^*(0, f+f_1+f_2)\tilde{U}_x(0, f+f_2)+\right. \\
& \left.\tilde{U}_x(0, f+f_1)\tilde{U}_y^*(0, f+f_1+f_2)\tilde{U}_y(0, f+f_2)]f_1f_2\cdot\right. \\
& \left.\int_{-\infty}^{\infty}\tilde{\eta}(f_3f_4)^*[\tilde{U}_x^*(0, u+f_3)\tilde{U}_x(0, u+f_3+f_4)\tilde{U}_x^*(0, u+f_4)+\right. \\
& \left.\tilde{U}_x^*(0, u+f_3)\tilde{U}_y(0, u+f_3+f_4)\tilde{U}_y^*(0, u+f_4)]f_3f_4\right] = \\
& \iiint_{-\infty}^{\infty}f_1f_2f_3f_4\tilde{\eta}(f_1f_2)\tilde{\eta}(f_3f_4)^*\cdot \\
& \{E[\tilde{U}_x(0, f+f_1)\tilde{U}_x^*(0, f+f_1+f_2)\tilde{U}_x(0, f+f_2)\tilde{U}_x^*(0, u+f_3)\tilde{U}_x(0, u+f_3+f_4)\tilde{U}_x^*(0, u+f_4)]+ \\
& E[\tilde{U}_x(0, f+f_1)\tilde{U}_y^*(0, f+f_1+f_2)\tilde{U}_y(0, f+f_2)\tilde{U}_x^*(0, u+f_3)\tilde{U}_y(0, u+f_3+f_4)\tilde{U}_y^*(0, u+f_4)]+ \\
& 2\text{Re}(E[\tilde{U}_x(0, f+f_1)\tilde{U}_x^*(0, f+f_1+f_2)\tilde{U}_x(0, f+f_2)\tilde{U}_x^*(0, u+f_3)\tilde{U}_y(0, u+f_3+f_4)\tilde{U}_y^*(0, u+f_4)])\}. \quad (12)
\end{aligned}$$

Now, putting together Theorems 1 and 2, Appendix 1 shows the following

Theorem 3

For jointly stationary circular complex Gaussian zero-mean processes $A(t), B(t), C(t), D(t), E(t), F(t)$ we have the general formula

$$\begin{aligned}
E\left[\tilde{A}(f+f_1)\tilde{B}^*(f+f_1+f_2)\tilde{C}(f+f_2)\tilde{D}^*(u+f_3)\tilde{E}(u+f_3+f_4)\tilde{F}^*(u+f_4)\right] = \\
\left[G_{ab}(f+f_1)G_{cd}(f)G_{ef}(f+f_4)\delta(f_2)\delta(f_3)+\right. \\
G_{ab}(f+f_1)G_{ed}(f+f_3)G_{cf}(f)\delta(f_2)\delta(f_4)+ \\
G_{cb}(f+f_2)G_{ad}(f)G_{ef}(f+f_4)\delta(f_1)\delta(f_3)+ \\
G_{cb}(f+f_2)G_{ed}(f+f_3)G_{af}(f)\delta(f_1)\delta(f_4)+ \\
G_{eb}(f+f_1+f_2)G_{ad}(f+f_1)G_{cf}(f+f_2)\delta(f_3-f_1)\delta(f_4-f_2)+ \\
\left.G_{eb}(f+f_1+f_2)G_{cd}(f+f_2)G_{af}(f+f_1)\delta(f_4-f_1)\delta(f_3-f_2)\right] \\
\cdot\delta(u-f) \quad \square \quad (13)
\end{aligned}$$

We next apply the general formula (13) to the three expectations in (12) to get:

First expectation:

$$\begin{aligned}
& E\left[\tilde{U}_x(0, f+f_1)\tilde{U}_x^*(0, f+f_1+f_2)\tilde{U}_x(0, f+f_2)\tilde{U}_x^*(0, u+f_3)\tilde{U}_x(0, u+f_3+f_4)\tilde{U}_x^*(0, u+f_4)\right] = \\
& \delta(u-f)\left[G_{xx}(f+f_1)G_{xx}(f)G_{xx}(f+f_4)\delta(f_2)\delta(f_3)+\right. \\
& G_{xx}(f+f_1)G_{xx}(f+f_3)G_{xx}(f)\delta(f_2)\delta(f_4)+ \\
& G_{xx}(f+f_2)G_{xx}(f)G_{xx}(f+f_4)\delta(f_1)\delta(f_3)+ \\
& G_{xx}(f+f_2)G_{xx}(f+f_3)G_{xx}(f)\delta(f_1)\delta(f_4)+ \\
& G_{xx}(f+f_1+f_2)G_{xx}(f+f_1)G_{xx}(f+f_2)\delta(f_3-f_1)\delta(f_4-f_2)+ \\
& \left.G_{xx}(f+f_1+f_2)G_{xx}(f+f_2)G_{xx}(f+f_1)\delta(f_4-f_1)\delta(f_3-f_2)\right] \quad (14)
\end{aligned}$$

where $G_{xx} \equiv \hat{G}_x$. Second expectation:

$$\begin{aligned}
& E \left[\tilde{U}_x(0, f + f_1) \tilde{U}_y^*(0, f + f_1 + f_2) \tilde{U}_y(0, f + f_2) \tilde{U}_x^*(0, u + f_3) \tilde{U}_y(0, u + f_3 + f_4) \tilde{U}_y^*(0, u + f_4) \right] = \\
& \delta(u - f) \left[G_{xy}(f + f_1) G_{yx}(f) G_{yy}(f + f_4) \delta(f_2) \delta(f_3) + \right. \\
& G_{xy}(f + f_1) G_{yx}(f + f_3) G_{yy}(f) \delta(f_2) \delta(f_4) + \\
& G_{yy}(f + f_2) G_{xx}(f) G_{yy}(f + f_4) \delta(f_1) \delta(f_3) + \\
& G_{yy}(f + f_2) G_{yx}(f + f_3) G_{xy}(f) \delta(f_1) \delta(f_4) + \\
& G_{yy}(f + f_1 + f_2) G_{xx}(f + f_1) G_{yy}(f + f_2) \delta(f_3 - f_1) \delta(f_4 - f_2) + \\
& \left. G_{yy}(f + f_1 + f_2) G_{yx}(f + f_2) G_{xy}(f + f_1) \delta(f_4 - f_1) \delta(f_3 - f_2) \right] \quad (15)
\end{aligned}$$

where $G_{yy} \equiv \hat{G}_y$, and assuming uncorrelated X and Y we get

$$\begin{aligned}
& E \left[\tilde{U}_x(0, f + f_1) \tilde{U}_y^*(0, f + f_1 + f_2) \tilde{U}_y(0, f + f_2) \tilde{U}_x^*(0, u + f_3) \tilde{U}_y(0, u + f_3 + f_4) \tilde{U}_y^*(0, u + f_4) \right] = \\
& \delta(u - f) \left[G_{yy}(f + f_2) G_{xx}(f) G_{yy}(f + f_4) \delta(f_1) \delta(f_3) + \right. \\
& \left. G_{yy}(f + f_1 + f_2) G_{xx}(f + f_1) G_{yy}(f + f_2) \delta(f_3 - f_1) \delta(f_4 - f_2) \right]. \quad (16)
\end{aligned}$$

Third expectation:

$$\begin{aligned}
& E \left[\tilde{U}_x(0, f + f_1) \tilde{U}_x^*(0, f + f_1 + f_2) \tilde{U}_x(0, f + f_2) \tilde{U}_x^*(0, u + f_3) \tilde{U}_y(0, u + f_3 + f_4) \tilde{U}_y^*(0, u + f_4) \right] = \\
& \delta(u - f) \left[G_{xx}(f + f_1) G_{xx}(f) G_{yy}(f + f_4) \delta(f_2) \delta(f_3) + \right. \\
& G_{xx}(f + f_1) G_{yx}(f + f_3) G_{xy}(f) \delta(f_2) \delta(f_4) + \\
& G_{xx}(f + f_2) G_{xx}(f) G_{yy}(f + f_4) \delta(f_1) \delta(f_3) + \\
& G_{xx}(f + f_2) G_{yx}(f + f_3) G_{xy}(f) \delta(f_1) \delta(f_4) + \\
& G_{yx}(f + f_1 + f_2) G_{xx}(f + f_1) G_{xy}(f + f_2) \delta(f_3 - f_1) \delta(f_4 - f_2) + \\
& \left. G_{yx}(f + f_1 + f_2) G_{xx}(f + f_2) G_{xy}(f + f_1) \delta(f_4 - f_1) \delta(f_3 - f_2) \right] \quad (17)
\end{aligned}$$

and assuming uncorrelated X and Y we get

$$\begin{aligned}
& E \left[\tilde{U}_x(0, f + f_1) \tilde{U}_x^*(0, f + f_1 + f_2) \tilde{U}_x(0, f + f_2) \tilde{U}_x^*(0, u + f_3) \tilde{U}_y(0, u + f_3 + f_4) \tilde{U}_y^*(0, u + f_4) \right] = \\
& \delta(u - f) \left[G_{xx}(f + f_1) G_{xx}(f) G_{yy}(f + f_4) \delta(f_2) \delta(f_3) + \right. \\
& \left. G_{xx}(f + f_2) G_{xx}(f) G_{yy}(f + f_4) \delta(f_1) \delta(f_3) \right]. \quad (18)
\end{aligned}$$

Substitution of (14),(16),(18) into (12) finally gives

$$\begin{aligned}
& \frac{E[\tilde{U}_{x,p}(L, f) \tilde{U}_{x,p}^*(L, u)]}{\Phi_{NL}^2} = \delta(u - f) \left\{ \iiint_{-\infty}^{\infty} \mathbf{f}_1 \mathbf{f}_2 \mathbf{f}_3 \mathbf{f}_4 \tilde{\eta}(f_1 f_2) \tilde{\eta}(f_3 f_4)^* \cdot \right. \\
& \cdot \left\{ \hat{G}_x(f + f_1) \hat{G}_x(f) \hat{G}_x(f + f_4) \delta(f_2) \delta(f_3) + \hat{G}_x(f + f_1) \hat{G}_x(f + f_3) \hat{G}_x(f) \delta(f_2) \delta(f_4) + \right. \\
& \hat{G}_x(f + f_2) \hat{G}_x(f) \hat{G}_x(f + f_4) \delta(f_1) \delta(f_3) + \hat{G}_x(f + f_2) \hat{G}_x(f + f_3) \hat{G}_x(f) \delta(f_1) \delta(f_4) + \\
& \hat{G}_x(f + f_1 + f_2) \hat{G}_x(f + f_1) \hat{G}_x(f + f_2) \delta(f_3 - f_1) \delta(f_4 - f_2) + \\
& \left. \hat{G}_x(f + f_1 + f_2) \hat{G}_x(f + f_2) \hat{G}_x(f + f_1) \delta(f_4 - f_1) \delta(f_3 - f_2) \right\} + \\
& \left[\hat{G}_y(f + f_2) \hat{G}_x(f) \hat{G}_y(f + f_4) \delta(f_1) \delta(f_3) + \hat{G}_y(f + f_1 + f_2) \hat{G}_x(f + f_1) \hat{G}_y(f + f_2) \delta(f_3 - f_1) \delta(f_4 - f_2) \right] + \\
& \left. 2\text{Re} \left[\hat{G}_x(f + f_1) \hat{G}_x(f) \hat{G}_y(f + f_4) \delta(f_2) \delta(f_3) + \hat{G}_x(f + f_2) \hat{G}_x(f) \hat{G}_y(f + f_4) \delta(f_1) \delta(f_3) \right] \right\}.
\end{aligned}$$

From (11), the term multiplying $\delta(u - f)$ must be the desired PSD. Each pair of delta removes two integrals, so that the PSD turns out to be (first two lines above produce first 4 lines, third line above produces 5th line, 4th line above produces 6th and 7th lines, and final line above produces the last two lines):

$$\begin{aligned}
\frac{\hat{G}_{x,p}(f)}{\Phi_{NL}^2} = & |\tilde{\eta}(0)|^2 \iint_{-\infty}^{\infty} \hat{G}_x(f+f_1) \hat{G}_x(f) \hat{G}_x(f+f_4) \mathfrak{f}_1 \mathfrak{f}_4 \\
& + |\tilde{\eta}(0)|^2 \iint_{-\infty}^{\infty} \hat{G}_x(f+f_1) \hat{G}_x(f+f_3) \hat{G}_x(f) \mathfrak{f}_1 \mathfrak{f}_3 \\
& + |\tilde{\eta}(0)|^2 \iint_{-\infty}^{\infty} \hat{G}_x(f+f_2) \hat{G}_x(f) \hat{G}_x(f+f_4) \mathfrak{f}_2 \mathfrak{f}_4 \\
& + |\tilde{\eta}(0)|^2 \iint_{-\infty}^{\infty} \hat{G}_x(f+f_2) \hat{G}_x(f+f_3) \hat{G}_x(f) \mathfrak{f}_2 \mathfrak{f}_3 \\
& + 2 \iint_{-\infty}^{\infty} |\tilde{\eta}(f_1 f_2)|^2 \hat{G}_x(f+f_1+f_2) \hat{G}_x(f+f_1) \hat{G}_x(f+f_2) \mathfrak{f}_1 \mathfrak{f}_2 \\
& + |\tilde{\eta}(0)|^2 \iint_{-\infty}^{\infty} \hat{G}_y(f+f_2) \hat{G}_x(f) \hat{G}_y(f+f_4) \mathfrak{f}_2 \mathfrak{f}_4 \\
& + \iint_{-\infty}^{\infty} |\tilde{\eta}(f_1 f_2)|^2 \hat{G}_y(f+f_1+f_2) \hat{G}_x(f+f_1) \hat{G}_y(f+f_2) \mathfrak{f}_1 \mathfrak{f}_2 \\
& + 2 \left(|\tilde{\eta}(0)|^2 \iint_{-\infty}^{\infty} \hat{G}_x(f+f_1) \hat{G}_x(f) \hat{G}_y(f+f_4) \mathfrak{f}_1 \mathfrak{f}_4 + \right. \\
& \quad \left. |\tilde{\eta}(0)|^2 \iint_{-\infty}^{\infty} \hat{G}_x(f+f_2) \hat{G}_x(f) \hat{G}_y(f+f_4) \mathfrak{f}_2 \mathfrak{f}_4 \right).
\end{aligned}$$

In summary, considering that by construction $\tilde{\eta}(0) = 1$, we have:

$$\begin{aligned}
\frac{\hat{G}_{x,p}(f)}{\Phi_{NL}^2} = & 2 \iint_{-\infty}^{\infty} |\tilde{\eta}(f_1 f_2)|^2 \hat{G}_x(f+f_1+f_2) \hat{G}_x(f+f_1) \hat{G}_x(f+f_2) \mathfrak{f}_1 \mathfrak{f}_2 \\
& + \iint_{-\infty}^{\infty} |\tilde{\eta}(f_1 f_2)|^2 \hat{G}_y(f+f_1+f_2) \hat{G}_x(f+f_1) \hat{G}_y(f+f_2) \mathfrak{f}_1 \mathfrak{f}_2 \\
& + \hat{G}_x(f) \left(4 \left(\int_{-\infty}^{\infty} \hat{G}_x(f) df \right)^2 + 4 \int_{-\infty}^{\infty} \hat{G}_x(f) df \int_{-\infty}^{\infty} \hat{G}_y(f) df + 1 \left(\int_{-\infty}^{\infty} \hat{G}_y(f) df \right)^2 \right)
\end{aligned}$$

which confirms Johannisson's equation (7) and completes the desired alternative proof.

V. CONCLUSIONS

We have presented an alternative derivation of Johannisson's result [1]. We first remark that our new method is able to deal with correlated X and Y, although this feature was not exploited in the present paper. Next we note that we did not have to assume independent input spectral lines: this comes naturally from the stationarity of the input process. Finally, the truly critical assumption in the model in [1], [2] is therefore the assumption of Gaussianity at any z during propagation, which is implicit in the assumption of a Gaussian input process, and the fact that the "forcing terms" in the RP equation are the linearly distorted signals at any z , which thus remain Gaussian.

Therefore the true limit of the model in [1], [2] is that indeed starting from a non-Gaussian spectrum such as the one of a digitally modulated signal¹, it takes some finite propagation in a non-infinite dispersion line to approximately get both a Gaussian spectrum and a Gaussian-like time-domain signal.

APPENDIX 1

In this Appendix we prove Theorem 3 in the text. Assuming jointly stationary circular complex Gaussian zero-mean processes $A(t)$, $B(t)$, $C(t)$, $D(t)$, $E(t)$, $F(t)$, we have by using (8) in (10):

$$\begin{aligned}
T \triangleq & E \left[\tilde{A}(f+f_1) \tilde{B}^*(f+f_1+f_2) \tilde{C}(f+f_2) \tilde{D}^*(u+f_3) \tilde{E}(u+f_3+f_4) \tilde{F}^*(u+f_4) \right] = \\
& \underbrace{E[\tilde{B}^*(f+f_1+f_2) \tilde{A}(f+f_1)]}_{G_{ab}(f+f_1)\delta(f_2)} \underbrace{E[\tilde{D}^*(u+f_3) \tilde{C}(f+f_2)]}_{G_{cd}(f+f_2)\delta(u+f_3-f-f_2)} \underbrace{E[\tilde{F}^*(u+f_4) \tilde{E}(u+f_3+f_4)]}_{G_{ef}(u+f_3+f_4)\delta(f_3)} + \\
& \underbrace{E[\tilde{B}^*(f+f_1+f_2) \tilde{A}(f+f_1)]}_{G_{ab}(f+f_1)\delta(f_2)} \underbrace{E[\tilde{D}^*(u+f_3) \tilde{E}(u+f_3+f_4)]}_{G_{ed}(u+f_3+f_4)\delta(f_4)} \underbrace{E[\tilde{F}^*(u+f_4) \tilde{C}(f+f_2)]}_{G_{cf}(f+f_2)\delta(u+f_4-f-f_2)} + \\
& \underbrace{E[\tilde{B}^*(f+f_1+f_2) \tilde{C}(f+f_2)]}_{G_{cb}(f+f_2)\delta(f_1)} \underbrace{E[\tilde{D}^*(u+f_3) \tilde{A}(f+f_1)]}_{G_{ad}(f+f_1)\delta(u+f_3-f-f_1)} \underbrace{E[\tilde{F}^*(u+f_4) \tilde{E}(u+f_3+f_4)]}_{G_{ef}(u+f_3+f_4)\delta(f_3)} + \\
& \underbrace{E[\tilde{B}^*(f+f_1+f_2) \tilde{C}(f+f_2)]}_{G_{cb}(f+f_2)\delta(f_1)} \underbrace{E[\tilde{D}^*(u+f_3) \tilde{E}(u+f_3+f_4)]}_{G_{ed}(u+f_3+f_4)\delta(f_4)} \underbrace{E[\tilde{F}^*(u+f_4) \tilde{A}(f+f_1)]}_{G_{af}(f+f_1)\delta(u+f_4-f-f_1)} + \\
& \underbrace{E[\tilde{B}^*(f+f_1+f_2) \tilde{E}(u+f_3+f_4)]}_{G_{eb}(u+f_3+f_4)\delta(f+f_1+f_2-u-f_3-f_4)} \underbrace{E[\tilde{D}^*(u+f_3) \tilde{A}(f+f_1)]}_{G_{ad}(f+f_1)\delta(u+f_3-f-f_1)} \underbrace{E[\tilde{F}^*(u+f_4) \tilde{C}(f+f_2)]}_{G_{cf}(f+f_2)\delta(u+f_4-f-f_2)} + \\
& \underbrace{E[\tilde{B}^*(f+f_1+f_2) \tilde{E}(u+f_3+f_4)]}_{G_{eb}(u+f_3+f_4)\delta(f+f_1+f_2-u-f_3-f_4)} \underbrace{E[\tilde{D}^*(u+f_3) \tilde{C}(f+f_2)]}_{G_{cd}(f+f_2)\delta(u+f_3-f-f_2)} \underbrace{E[\tilde{F}^*(u+f_4) \tilde{A}(f+f_1)]}_{G_{af}(f+f_1)\delta(u+f_4-f-f_1)} =
\end{aligned}$$

thus

$$\begin{aligned}
T = & G_{ab}(f+f_1)G_{cd}(f+f_2)G_{ef}(u+f_3+f_4)\delta(f_2)\delta(f_3)\delta(u+f_3-f-f_2) \\
& + G_{ab}(f+f_1)G_{ed}(u+f_3+f_4)G_{cf}(f+f_2)\delta(f_2)\delta(f_4)\delta(u+f_4-f-f_2) \\
& + G_{cb}(f+f_2)G_{ad}(f+f_1)G_{ef}(u+f_3+f_4)\delta(f_1)\delta(f_3)\delta(u+f_3-f-f_1) \\
& + G_{cb}(f+f_2)G_{ed}(u+f_3+f_4)G_{af}(f+f_1)\delta(f_1)\delta(f_4)\delta(u+f_4-f-f_1) \\
& + G_{eb}(u+f_3+f_4)G_{ad}(f+f_1)G_{cf}(f+f_2) \cdot \\
& \cdot \delta(u+f_4-f-f_2)\delta(u+f_3-f-f_1)\delta(f+f_1+f_2-u-f_3-f_4) \\
& + G_{eb}(u+f_3+f_4)G_{cd}(f+f_2)G_{af}(f+f_1) \cdot \\
& \cdot \delta(u+f_4-f-f_1)\delta(u+f_3-f-f_2)\delta(f+f_1+f_2-u-f_3-f_4).
\end{aligned}$$

Now we use the sampling property of the delta to write, e.g. for the first line where $f_2 = 0$ and $f_3 = 0$,

$$G_{ab}(f+f_1)G_{cd}(f)G_{ef}(u+f_4)\delta(f_2)\delta(f_3)\delta(u-f)$$

and e.g. for the last line where $u+f_4 = f+f_1$ and $u+f_3 = f+f_2$ which we add up to get

$$u+f_3+f_4 = (f-u) + f+f_1+f_2$$

whence

$$f+f_1+f_2-u-f_3-f_4 = u-f$$

so that the last line writes as

$$\begin{aligned}
& G_{eb}(u+f_3+f_4)G_{cd}(f+f_2)G_{af}(f+f_1)\delta(u+f_4-f-f_1)\delta(u+f_3-f-f_2)\delta(f+f_1+f_2-u-f_3-f_4) = \\
& G_{eb}((f-u)+f+f_1+f_2)G_{cd}(f+f_2)G_{af}(f+f_1) \cdot \delta(u+f_4-f-f_1)\delta(u+f_3-f-f_2)\delta(u-f) \stackrel{(\text{use } u=f)}{=} \\
& G_{eb}(f+f_1+f_2)G_{cd}(f+f_2)G_{af}(f+f_1)\delta(f_4-f_1)\delta(f_3-f_2)\delta(u-f).
\end{aligned}$$

¹Although the authors in [2] present in their Appendix B an appealing heuristic justification of their Gaussian signal assumption, still their invoking the central limit theorem at their equation (37) is *not rigorous*. They would conclude that any digitally modulated signal with any number of levels has a Gaussian Fourier transform (which in turn implies the time-domain signal itself is Gaussian), which is clearly *not* the case.

We therefore get

$$\begin{aligned}
T = & G_{ab}(f + f_1)G_{cd}(f)G_{ef}(f + f_4)\delta(f_2)\delta(f_3)\delta(u - f) \\
& + G_{ab}(f + f_1)G_{ed}(f + f_3)G_{cf}(f)\delta(f_2)\delta(f_4)\delta(u - f) \\
& + G_{cb}(f + f_2)G_{ad}(f)G_{ef}(f + f_4)\delta(f_1)\delta(f_3)\delta(u - f) \\
& + G_{cb}(f + f_2)G_{ed}(f + f_3)G_{af}(f)\delta(f_1)\delta(f_4)\delta(u - f) \\
& + G_{eb}(f + f_1 + f_2)G_{ad}(f + f_1)G_{cf}(f + f_2)\delta(f_4 - f_2)\delta(f_3 - f_1)\delta(u - f) \\
& + G_{eb}(f + f_1 + f_2)G_{cd}(f + f_2)G_{af}(f + f_1)\delta(f_4 - f_1)\delta(f_3 - f_2)\delta(u - f).
\end{aligned}$$

whence the final form (13) given in Theorem 3.